ECE 440, HW#9, Kefu Zhu

Question 7

(A)

Because $t_1 \neq t_2 \rightarrow t_1 - t_2 \neq 0$, and we have

$$\delta(t) = \begin{cases} \infty, \ t = 0 \\ 0, \ t \neq 0 \end{cases}$$

so $\delta(t_1-t_2)=0$. Then $R_W(t_1,t_2)=0$, which tells us that $W(t_1)$ and $W(t_2)$ are uncorrelated. But since W(t) is a Gaussian process, uncorrelation implies independence, so $W(t_1)$ and $W(t_2)$ are also independent.

(B)

(1) Show that the process X(t) is Gaussian

Because W(t) is a Gaussian process and X(t) is a linear functional of W(t), X(t) is also a Gaussian process.

(2) Compute the mean and autocorrelation functions of $\boldsymbol{X}(t)$

$$\mu_X(t) = E[\int_0^t W(u) \, du] = \int_0^t E[W(u)] du = \int_0^t \mu_W(u) du = 0$$

$$R_X(t_1, t_2) = E[\int_0^{t_1} \int_0^{t_2} W(u)W(v) dv du]$$

$$= \int_0^{t_1} \int_0^{t_2} E[W(u)W(v)] dv du$$

$$= \int_0^{t_1} \int_0^{t_2} \sigma^2 \cdot \delta(u - v) \, dv \, du$$

• If
$$t_1 < t_2$$
:

$$= \int_0^{t_1} \int_0^{t_1} \sigma^2 \cdot \delta(u - v) \, dv \, du + \int_0^{t_1} \int_{t_1}^{t_2} \sigma^2 \cdot \delta(u - v) \, dv \, du$$

$$= \int_0^{t_1} \int_0^{t_1} \sigma^2 \cdot \delta(u - v) \, dv \, du$$

$$=\int_0^{t_1} \sigma^2 du$$

$$= \sigma^2 \cdot t_1$$

• If $t_1 > t_2$:

$$= \int_0^{t_2} \int_0^{t_2} \sigma^2 \cdot \delta(u - v) \, dv \, du + \int_0^{t_2} \int_{t_2}^{t_1} \sigma^2 \cdot \delta(u - v) \, dv \, du$$

$$= \int_0^{t_2} \int_0^{t_2} \sigma^2 \cdot \delta(u - v) \, dv \, du$$

$$=\int_0^{t_2}\sigma^2du$$

$$= \sigma^2 \cdot t_2$$

To summarize, $R_X(t_1, t_2) = \sigma^2 \cdot \min(t_1, t_2)$

(3) Compute P(X(t) > a) for arbitrary a and t > 0

$$var(X(t)) = E[X^{2}(t)] - E^{2}[X(t)] = R_{X}(t, t) - \mu_{X}^{2}(t) = \sigma^{2}t$$

Based on Gaussian pdf,

$$P(X(t) > a) = \int_0^\infty \frac{1}{\sqrt{(2\pi\sigma^2 t)}} \cdot \exp(-\frac{x^2}{2\sigma^2 t}) dx$$

(C)

$$\mu_{W_h}(n) = E[W_h(n)] = E[\int_{nh}^{(n+1)h} W(u) \ du] = \int_{nh}^{(n+1)h} E[W(u)] \ du = \int_{nh}^{(n+1)h} \mu_W(u) \ du = 0$$

$$R_{W_h}(n_1, n_2) = E\left[\int_{n_1 h}^{(n_1+1)h} \int_{n_2 h}^{(n_2+1)h} W(u)W(v) \ du \ dv\right]$$

$$= \int_{n_1h}^{(n_1+1)h} \int_{n_2h}^{(n_2+1)h} E[W(u)W(v)] \ du \ dv$$

$$= \int_{n_1h}^{(n_1+1)h} \int_{n_2h}^{(n_2+1)h} \sigma^2 \delta(u-v) \ du \ dv$$

- When $n_1 = n_2$, $R_{W_h}(n_1, n_2) = \sigma^2 h$
- When $n_1 \neq n_2$, $R_{W_h}(n_1, n_2) = 0$

By defining discrete-time $\delta_d(n)=\left\{ egin{array}{ll} 1, & n=0 \\ 0, & n
eq 0 \end{array} \right.$, we can express $R_{W_h}(n_1,n_2)$ as $\sigma^2h\delta_d(n_1-n_2)$

(D)

problem7_d

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% Set parameters
h=0.01; sigma=1; t_MAX=10;

W_vector=normrnd(0,sigma*sqrt(h),1,t_MAX/h);
X_vector=cumsum(W_vector);

% Plot sample path
plot(h:h:t_MAX,X_vector);
xlabel('time');title('Weiner Process Simulated, h = ' + string(h));
grid on; axis([0 t_MAX -5 5])
```

